$\mathrm{CS}\ 435$: Linear Optimization

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Lecture 19: Using LP techniques to design algorithms for combinatorial problems

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1 Integer Linear Programs

An *Integer Linear Program* is a linear program where the variables are constrained to take integer values only. Note that all variables have to take only integer values.

An ILP can be written as

$$\max_{\substack{Ax \leq b \\ x_i \text{ is integral } \forall i.}} \sum_{k=1}^{\infty} \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} \sum_{k=1}^{\infty} \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} \sum_{j=1}^{\infty} \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} \sum_{j=1}^{\infty} \sum_{i=1}^{\infty} \sum_{j=1}^{\infty} \sum_{j=1}^{\infty$$

The goal of this lecture is to present a technique for designing algorithms. We will give a template for designing *combinatorial* algorithms. The added advantage is that proving correctness comes for free. The template will come from linear programming; however the algorithms we design will be *combinatorial*.

To understand the template, we will try designing an iterative algorithm for the linear programming problem:

$$\max c^T x$$
$$Ax \le \mathbf{b} \tag{1}$$

The first step is to find some feasible solution. While this, in general, is as hard as solving an LP, in our examples, this will be easy. For instance, if $\mathbf{b} \ge \mathbf{0}$, then $x = \mathbf{0}$ will work.

We will design an iterative step that will be repeatedly executed till we reach an optimum. Suppose we have a feasible x^i after the *i*th step. In the i + 1th step, we will find a x^{i+1} which is also feasible, with increased cost.

At the beginning of the i + 1th iteration we have an x^i such that $Ax^i \leq \mathbf{b}$. We need an x^{i+1} such that $Ax^{i+1} \leq \mathbf{b}$ and $c^T(x^{i+1} - x^i) > 0$.

In other words, we need to figure out a direction vector y_i which is $x^{i+1} - x^i$. The cost must increase in this direction and moving in this direction should not land us outside the feasible region. How do we handle the second condition? It is clear that the inequalities which are equalities at x_i will play a role. For the strict inequalities, moving around x_{i+1} in a small enough neighbourhood will not lead to violations.

Let $A'x^i = \mathbf{b}'$; $A''x^i < \mathbf{b}''$. For feasibility we need a y_i such that $A'(x^i + \epsilon y_i) \le b'$ for some small $\epsilon > 0$. Simplifying, we see that we need a y_i such that $c^T y_i > 0$ and $A' y_i \le 0$.

Our objective now is: Design an algorithm to find such a y_i and update x^i suitably. Once we do this, we put this piece of code in a loop and we are done. This piece of code will be combinatorial and will depend on the problem at hand. Notice that this may be easier since the RHS is zero.

Exercise: Given such a y_i how do you find x^{i+1} ?

Exercise: Prove that if x^i is not optimal, such a y_i exists.

Hint: Assume that x_0 is the optimal point. Now consider the vector $\epsilon(x_0 - x^i)$. Note that this exercise provides a proof of correctness.

The final template is given below.

- 1. Find an initial feasible point x^0 .
- 2. Loop as long as you can.

{ In this loop we update x^i to x^{i+1} of increased cost. If we cannot, then the algorithm terminates.} Identify the equalities. That is, determine A' such that $A'x^i = b'$; $A''x^i < b''$. Find y_i satisfying

$$c^T y_i = 1; A' y_i \leq \mathbf{0}$$

{ The previous step will be replaced by a combinatorial algorithm to compute y_i } Determine the maximum $\epsilon > 0$ such that $x^i + \epsilon y_i$ is feasible. How? $x^{i+1} = x^i + \epsilon y_i$.